

# Stanford Machine Learning Week 2. Linear Regression with multiple variables.

- multiple inputs (information) to predict.
- input:  $x_1, x_2, \dots, x_n$ ; output =  $y$ .

e.g.

$x_1$ size	$x_2$ # of bedrooms	$x_3$ Number of floors	$x_4$ Age of home	price ( $y$ )
				m samples

Notation:

$n$  - number of features.

$x^{(i)}$  - input (features) of  $i$ th training sample

$x_j^{(i)}$  - value of feature  $j$  in  $i$ th training sample

$$\Rightarrow x^{(i)} = [x_j^{(i)}].$$

index to training set

$$x^{(i)} = \begin{pmatrix} x^{(i)}_1 \\ x^{(i)}_2 \\ \vdots \\ x^{(i)}_n \end{pmatrix}$$

$n$ -dimensional vector.

## - Hypothesis

previously  $h_\theta(x) = \theta_0 + \theta_1 x$

now  $\rightarrow h_\theta(x) = \theta_0 + \theta_1 x_1 + \dots + \theta_n x_n$

define:  $x_0 = 1$ , i.e.  $x_0^{(i)} = 1 \quad \forall i$

$$\Rightarrow x = \begin{bmatrix} x_0 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^{n+1}; \quad \theta = \begin{bmatrix} \theta_0 \\ \vdots \\ \theta_n \end{bmatrix} \in \mathbb{R}^n.$$

$$h_\theta(x) = \theta_0 x_0 + \dots + \theta_n x_n$$

$$= \theta^T \cdot x \quad (\text{expressed as inner product of vectors})$$

$$\Rightarrow \text{Multivariate Linear Regression: } h_\theta(x) = \theta^T \cdot x.$$